# Stability of the numerical scheme used for pricing green bonds

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- Option Pricing Theory
- Sustainable Finance
- Numerical Analysis
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- Research Questions





### Definition: Zero Coupon Bond (ZCB)

A zero-coupon bond with a value at time t with maturity time T, denoted by B(t, T), is a financial instrument that can be bought at time t = 0 for a price of B(0, T), and pays one unit of currency (euro, dollar, etc.) at maturity time T, i.e., B(T, T) = 1.



Figure: Payments for a zero-coupon bond with maturity time T.



• Constant interest *r*:

$$e^{r(T-t)}B(t,T) = 1 \Longrightarrow B(t,T) = e^{-r(T-t)}.$$



• Constant interest r:

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• Continuously differentiable interest:  $r : [0, T] \rightarrow \mathbb{R}$ :

$$B(t,T)=e^{-\int_t^T r(s)ds}.$$



### Definition: European call option

A European call option gives an option holder the right, but not the obligation, to buy an asset at a pre-specified maturity time T for a pre-specified strike price K from the option writer.



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Figure: The pay-off diagram of a European call option:  $(S_T - K)^+ := \max \{S_T - K, 0\}.$ 



### Definition: European put option

A European put option gives an option holder the right, but not the obligation, to sell an asset at a pre-specified maturity time T for a pre-specified strike price K from the option writer.



### Definition: European put option

A European put option gives an option holder the right, but not the obligation, to sell an asset at a pre-specified maturity time T for a pre-specified strike price K from the option writer.



Figure: The pay-off diagram of a European put option:  $(K - S_T)^+ := \max \{K - S(T), 0\}.$ 



- Value of a call option:  $V_{call}$ .
- Value of a put option:  $V_{put}$ .



- Value of a call option:  $V_{call}$ .
- Value of a put option: V<sub>put</sub>.

#### Theorem: Put-call parity

Let S be some asset,  $V_i(S, t)$  the value of an option at time t for  $i \in \{\text{call, put}\}$  with S being the underlying asset, K the strike price and T - t the time to maturity. Moreover, let r be the risk-free interest rate and assume that r is constant. If S(t) is the price of the asset at time t, we have the following equality:

$$V_{\mathsf{call}}(t,S) + Ke^{-r(T-t)} = V_{\mathsf{put}}(t,S) + S(t).$$



Black-Scholes model



Figure: From left to right: Robert Merton, Myron Scholes, and Fischer Black.



$$\begin{cases} \frac{\partial V}{\partial t} + \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} + rS \frac{\partial V}{\partial S} - rV = 0, \\ \forall S \ge 0 : V(T, S) = F(S). \end{cases}$$



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Parameter analysis:

- V: the value of the call or put option,
- S: the value of the asset (or stock),
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Boundary condition:

- For a call option:  $F(S) = (S_T K)^+$ .
- For a put option:  $F(S) = (K S_T)^+$ .



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$$d_1 = \frac{\log\left(\frac{S_0}{K}\right) + \left(r + \frac{1}{2}\sigma^2\right)T}{\sigma\sqrt{T}}$$

and

$$d_2 = \frac{\log\left(\frac{S_0}{K}\right) + \left(r - \frac{1}{2}\sigma^2\right)T}{\sigma\sqrt{T}} = d_1 - \sigma\sqrt{T}.$$

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Call option:

$$V_{\mathsf{call}}(t,S) = S(t)\Phi(d_1) - Ke^{-r(T-t)}\Phi(d_2).$$

• Put option:

$$V_{put}(t,S) = Ke^{-r(T-t)}\Phi(-d_2) - S(t)\Phi(-d_1).$$



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- Advantages:
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- Advantages:
  - Quick
  - Fair price for writer and holder of an option.
- Disadvantage:
  - r and  $\sigma$  are assumed to be constant.
  - Trading is not a continuous process.
  - Dividend payments are considered absent.



Brown derivatives —> Green derivatives



- Brown derivatives → Green derivatives
- Copying behavior





Green bonds

• Multiple definitions



- Multiple definitions
- Greenwashing



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- European Green Bond Standard (EUGBS)



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#### Key question: How can we determine the value of a green bond?



• Exact solution  $\rightarrow$  numerical solution (approximated).

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Figure 1: Finite difference grid  $\{mh, nk\}_{m,n}$  for  $0 \le m \le M$  and  $0 \le n \le N$ .



Time stepping methods:



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- Explicit
  - Forward Euler (FE)



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Differences between explicit and implicit time methods:

- Less computations are required for explicit methods.
- Implicit methods tend to be more stable than explicit methods.



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Subtracting (1) from (2), we obtain

$$\begin{cases} \epsilon' = \tilde{y}' - y' = \lambda \left( \tilde{y} - y \right) = \lambda \epsilon, t > t_0, \\ \epsilon \left( t_0 \right) = \epsilon_0, \end{cases}$$
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$$|Q(\lambda \Delta t)| \le 1 \iff$$
 Stability.  
 $|Q(\lambda \Delta t)| < 1 \iff$  Absolute stability.



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• Backward Euler and Crank Nicolson: unconditionally stable. We also will consider von Neumann stability.

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'Too expensive'

Iterative solution methods

$$\{\mathbf{u}^k\}_{k\geq 0}$$
 s.t.  $\lim_{k\to\infty}\mathbf{u}^k = \mathbf{u}.$ 



Three iterative solution methods:

- Conjugate Gradient
  - A must be positive definite:  $\forall \mathbf{v} \in \mathbb{R}^n \setminus \{\mathbf{0}\} : \mathbf{v}^\top A \mathbf{v} > 0$ ,
  - Short recurrences.



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  - Short recurrences.
- BiCG-STAB
  - Can be used for general matrices A,
  - Short recurrences,
  - No optimality property.
- GMRES
  - Can be used for general matrices A,
  - Long recurrences,
  - Optimality property.



The PDE for pricing a green bond is derived by Juriaan Rutten:

$$\begin{split} \frac{\partial V}{\partial t} &+ \left(\mu c - \lambda_c \sigma_c c\right) \frac{\partial V}{\partial c} + \left(\alpha (\beta - r) - \lambda_r \sigma_r \sqrt{r}\right) \frac{\partial V}{\partial r} \\ &+ \frac{1}{2} \left(\sigma_c^2 c^2 \frac{\partial^2 V}{\partial c^2} + \sigma_r^2 r \frac{\partial^2 V}{\partial r^2} + 2c\rho\sigma_c\sigma_r \sqrt{r} \frac{\partial^2 V}{\partial r\partial c}\right) - rV = 0. \end{split}$$

- c: carbon price,
- r: risk-free interest rate,
- $\sigma_c, \sigma_r$ : volatility of the carbon price and the risk-free rate.



### **Research Questions**

# Are the stability conditions for FE, BE, and CN still valid for this $\ensuremath{\mathsf{PDE}}\xspace$



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• How do the changing interest rate r and the volatility  $\sigma$  influence the stability of the numerical methods?



Are the stability conditions for FE, BE, and CN still valid for this PDE?

- How do the changing interest rate r and the volatility  $\sigma$  influence the stability of the numerical methods?
- What role do the boundary conditions play in when determining the stability?



Thank you for listening. Are there any questions?

